their	
simulation, origins	
scientists and	
to	
engineers embarked on another path: one in which is nature itself spectacular	wh

beginning of the

Genetic Algorithms Evolutionar

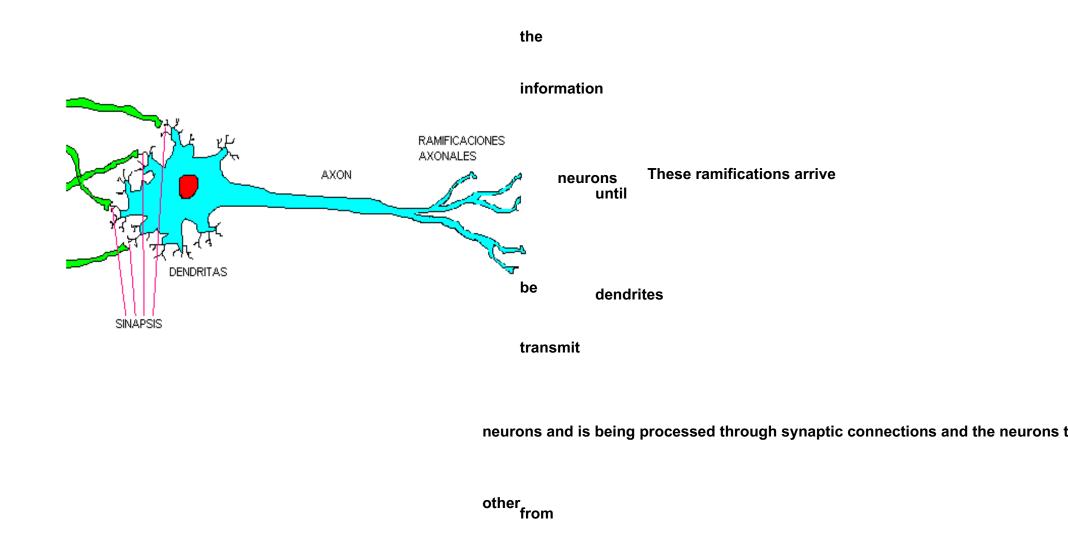
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human brain	consists	of millions	NEURAL of interco	NETWOI projected	RKS (c	OI 1S
human brain	consists	of millions	of interco	nnected	neuro	

learning models and approach inspired by the

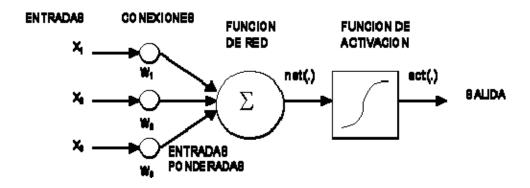
neuron

collects signals from other neurons via structures called dendrites. Neuron delive



some

other



function

synapse

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Activation f ⊮ாopiega∯ coutpuSign Weight	Synaptic Artificial
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Hebb

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the

connections the from connections be

in

neurons It represents

1949

neurons is strengthened if both are activated.

between two

simultaneously

in fundamental

two neurons

Dartmouth IA. Here the first simul publishes a

theory

of neural adaptation
and patterns inspired
by this theory,
Adaline (Adaptive
Linear Neuron) and
Madaline (Multiple Adaline). These models were used in numeron

critique of		of the		
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		learning ••	optical	
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		they e	extended	
	to	o use contin	uous signals inp	out and output. published
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Perceptron, revealing serfels limitations. This work created serious doub

brain-state	e-in StHoons (@6 ,B).
1984) continues the work of	Anderson and develops competitive
	Ritz
how	&
	Lancara
model	Jomnes
	1984) continues the work of a

(Nononen

from solves the problems posed by Minsky and Paper.
In this decade, the reviva illustrative East group on highlights which the mechanisms algorithm and retrieval of memory.

backpropagation, storage

remarkable

(Hinton

1986) and BAM models (Kos

are

& from

Sejnowski is

decade

They

de will be	design a network must establish: Structure of ar	
Will be		
units		

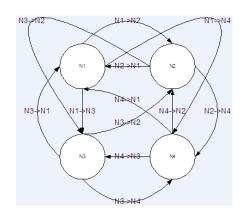
	sigmoidal activation radial
	basis functions
·	Dasis functions

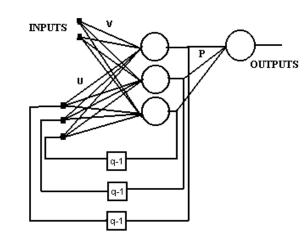
organized in layers
so that the output of a layer constitutes the input of the

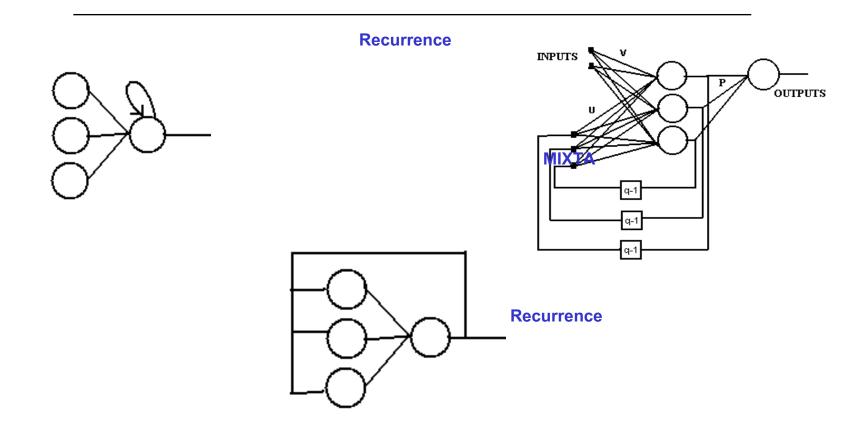
They have some type of feedback.

RNR ELMAN

Hopfield, SOM, Elman RECURRING SECOND







the input-output
rolationahin of
relationship of
the network captures the information in a table (X,

or approaches

Ве

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as its premise the biological plausibility of the models.

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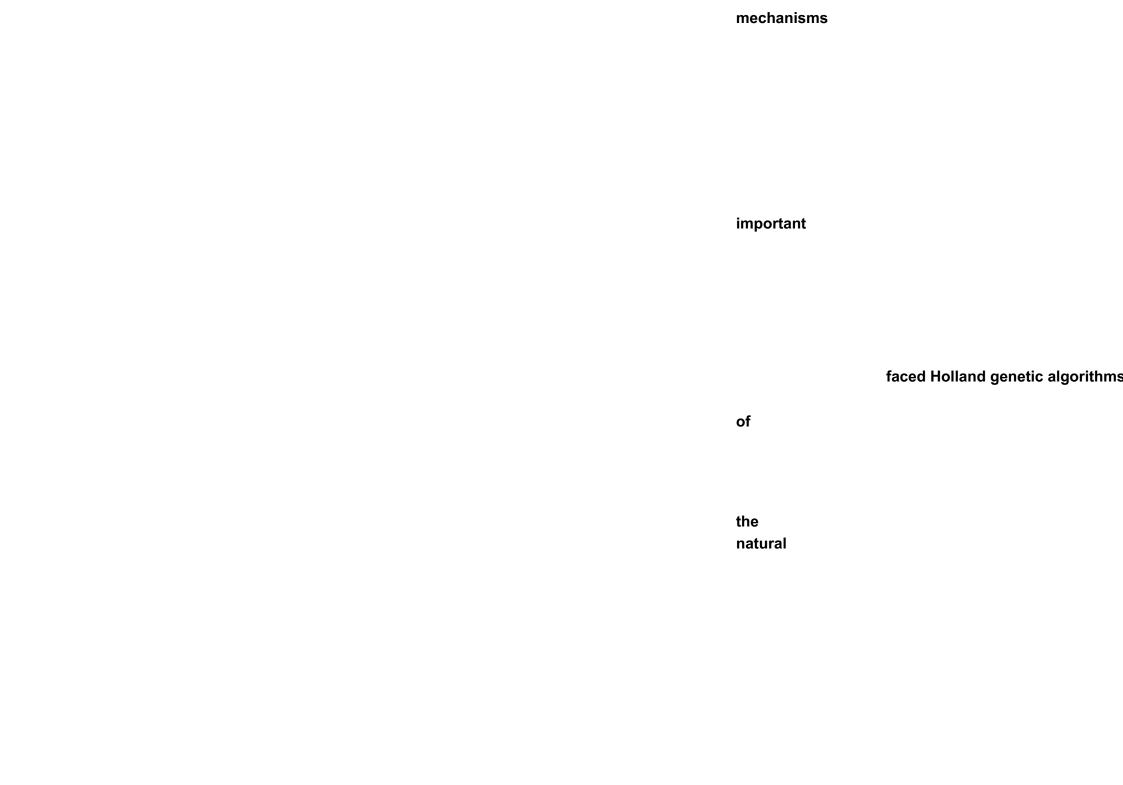
wno

the theories, inherited discovered that necessary mechanism for evolution.
they **provided** they forgot and did not turn to rediscover until the Mendel, who that were the isolation, job that characters the the in Mendelian total They

of deoxyribonucleic acid. the genes is about the same time, the in DNA,

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John Holland at



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we speak generically of Evolutionary Algorithms. Evolutionary prograr					

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Automata, (AC), it is a formal model con cells, entities or agents,

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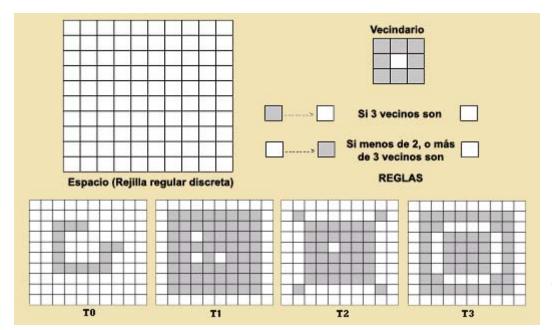
be number of cells is finite, the edges. The cells have n**dimeigation**rs edges beyond the limits of the reticle.

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Cellular automaton "Game of Life".

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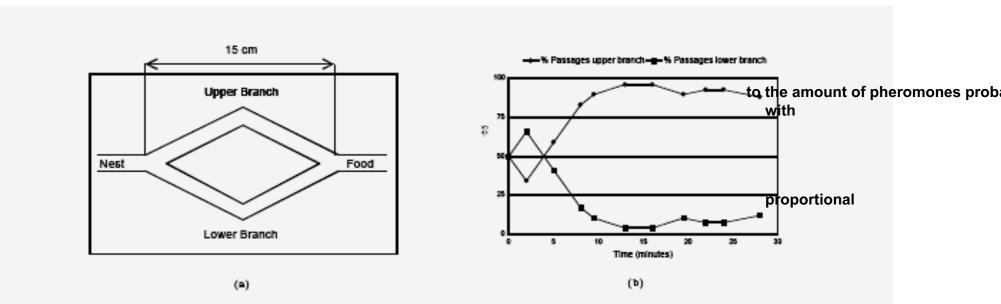
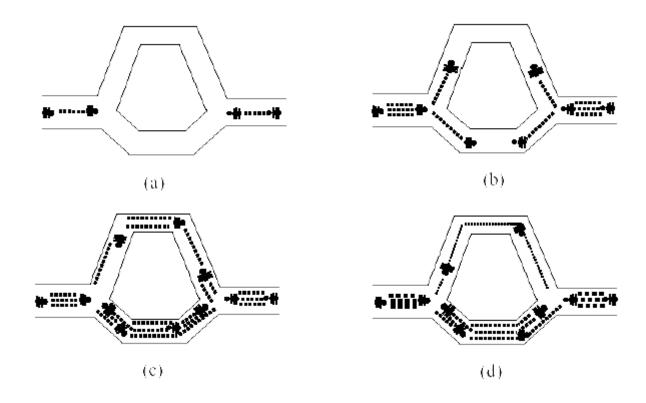


Figure 1. Single bridge experiment. (a) Experimental setup. (b) Results for a typical single trial, showing the percentage of passages on each of the two branches per unit of time as a function of time. Eventually, after an initial short transitory phase, the upper branch becomes the most used. After Deneubourg et al., 1990 [25].



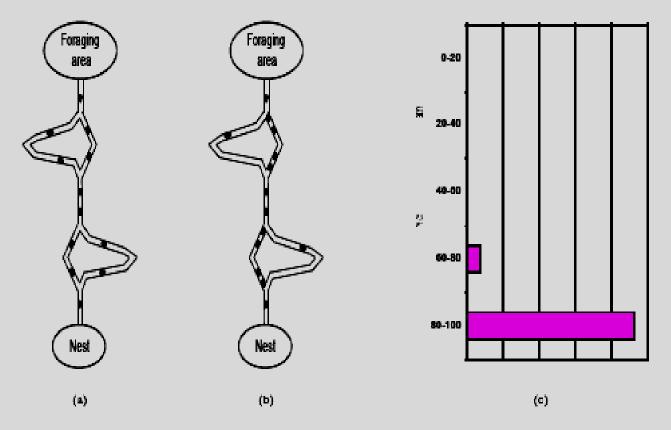


Figure 2. Double bridge experiment. (a) Ants start exploring the double bridge. (b) Eventually most of the ants choose the shortest path. (c) Distribution of the percentage of ants that selected the shorter path. After Goss et al. 1989 [60].

TSP). traveling salesman (Traveling Salesm	based on very simple agents on Problem
TSP). three	ian Frobieni

variants

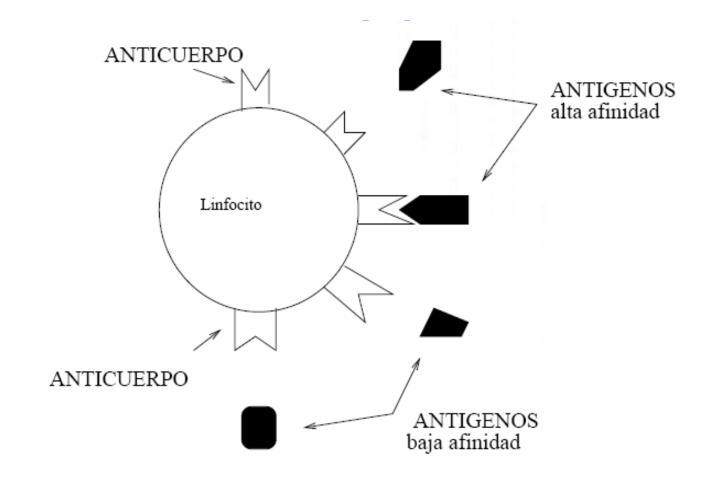
algorithm

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all these methods are infoluded within what is known as swarm intelligence.

Problem

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Detection

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Security in computer systems and networks _

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means 10 ^ 10 times more coeffision by, Data can be sto at run an approximate density of 21 bit per cubic nanometer (nm

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